

Solving Problems by Searching

CS 486/686: Introduction to Artificial Intelligence
Fall 2013

Outline

- Problem solving agents and search
- Examples
- Properties of search algorithms
- Uniformed search
 - Breadth first
 - Depth first
 - Iterative deepening

Introduction

- Search was one of the first topics studied in AI
 - Newell and Simon (1961) *General Problem Solver*
- Central component to many AI systems
 - Automated reasoning, theorem proving, robot navigation, scheduling, game playing,...

Defining a Search Problem

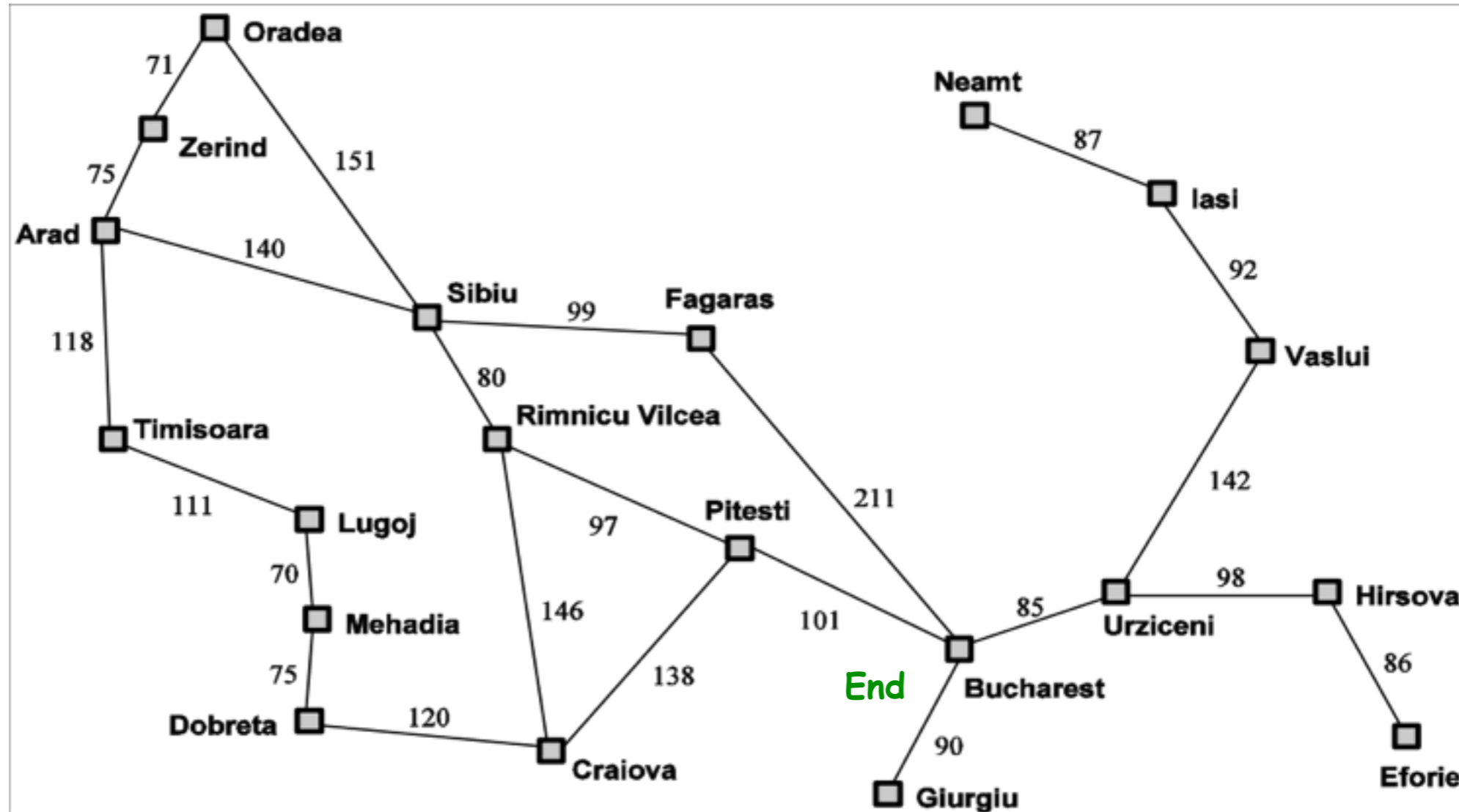
- **State space S** : all possible configs. of the domain
- **Initial state $s_0 \in S$** : the start state
- **Goal states $G \subseteq S$** : the set of end states
 - **Goal test**: check if we are in a goal state
- **Operators A** : actions available
 - Often defined in terms of **mappings** from **state** to **successor state**

Defining a Search Problem

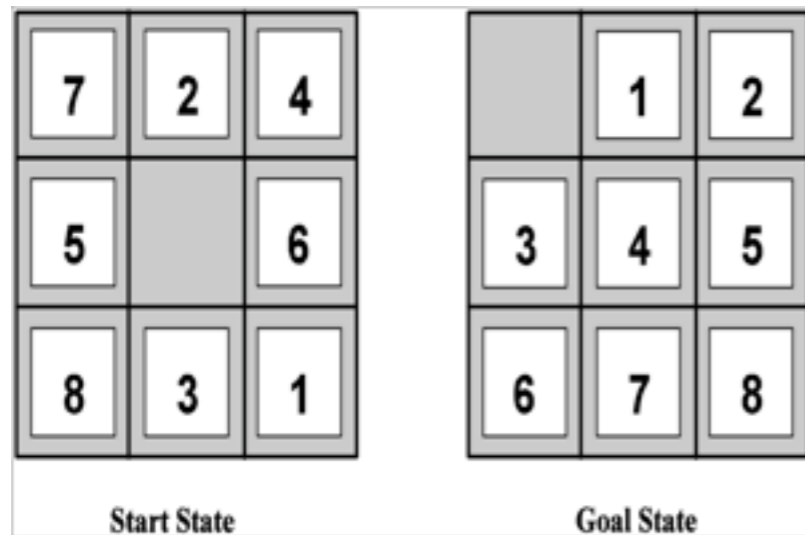
- **Path:** a sequence of states and operators
- **Path cost c :** a number associated with any path
- **Solution:** a path from s_0 to $s_G \in G$
- **Optimal solution:** a path with minimum cost

Example: Traveling in Romania

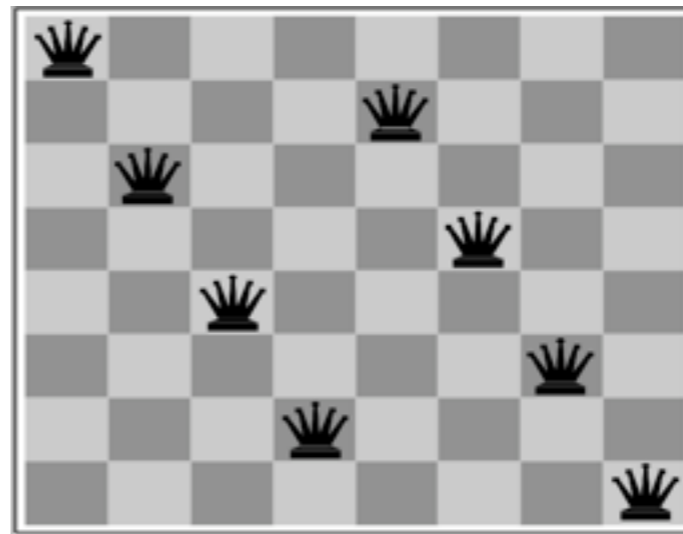
Start



Examples of Search Problems



- States:
- Initial State:
- Operators:
- Goal test:
- Path cost:



- States:
- Initial State:
- Operators:
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- Path cost:

Examples of Search Problems



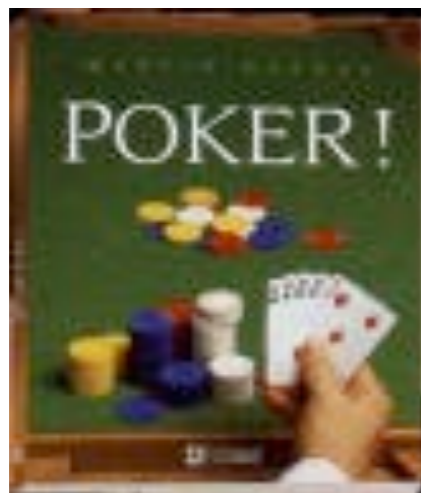
Our Definition Excludes...

Chance



Adversarial

Continuous states



Partial
Observability



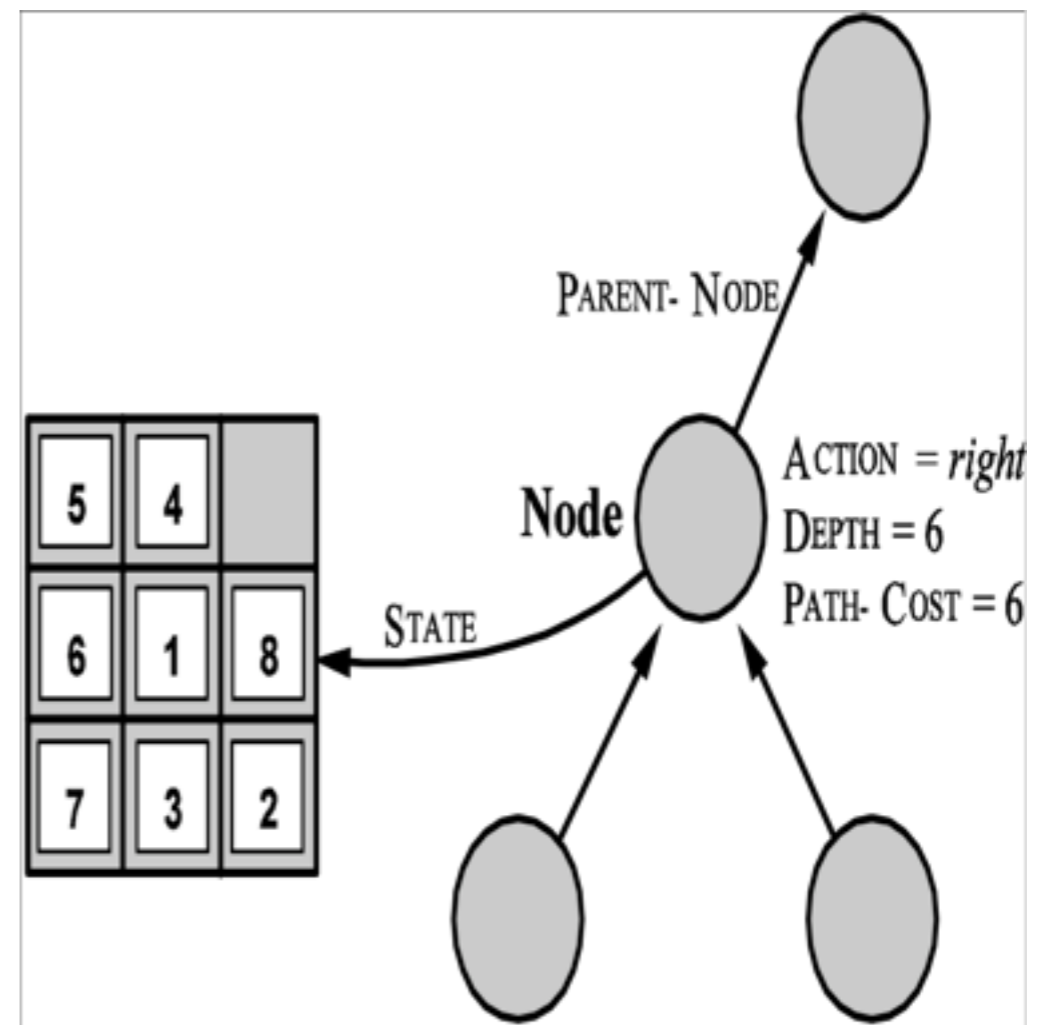
All of the above

Representing Search

- Search graph
 - Vertices correspond to states
 - Edges correspond to operators
- We search for a solution by building a search tree and traversing it to find a goal state

Data Structures: Search Node

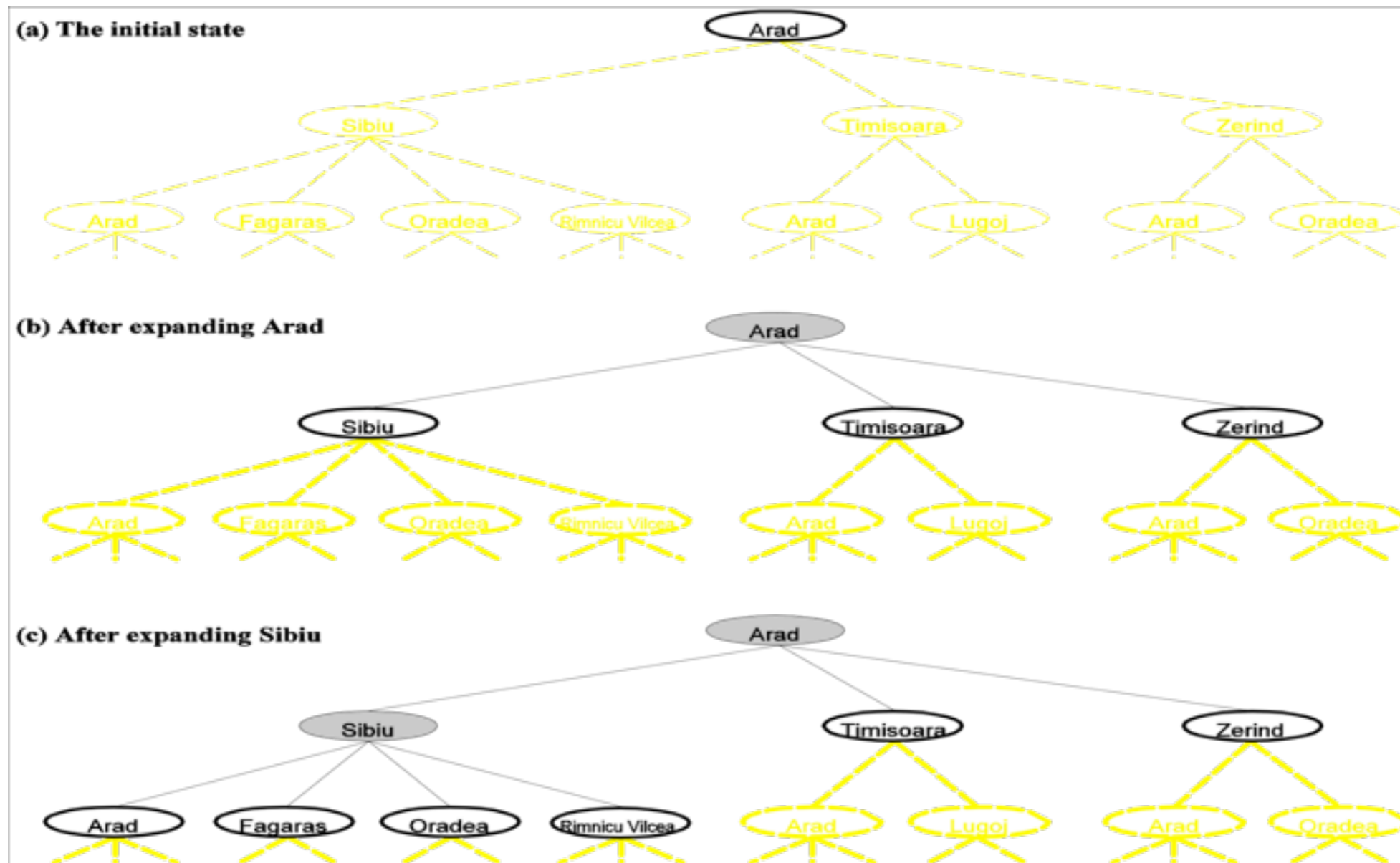
- State
- Parent node and operator applied to parent to reach current node
- Cost of path so far
- Depth of node



Expanding Nodes

- Expanding a node
 - Applying all legal operators to the state contained in the node
 - Generating nodes for all corresponding successor states

Expanding Nodes



Generic Search Algorithm

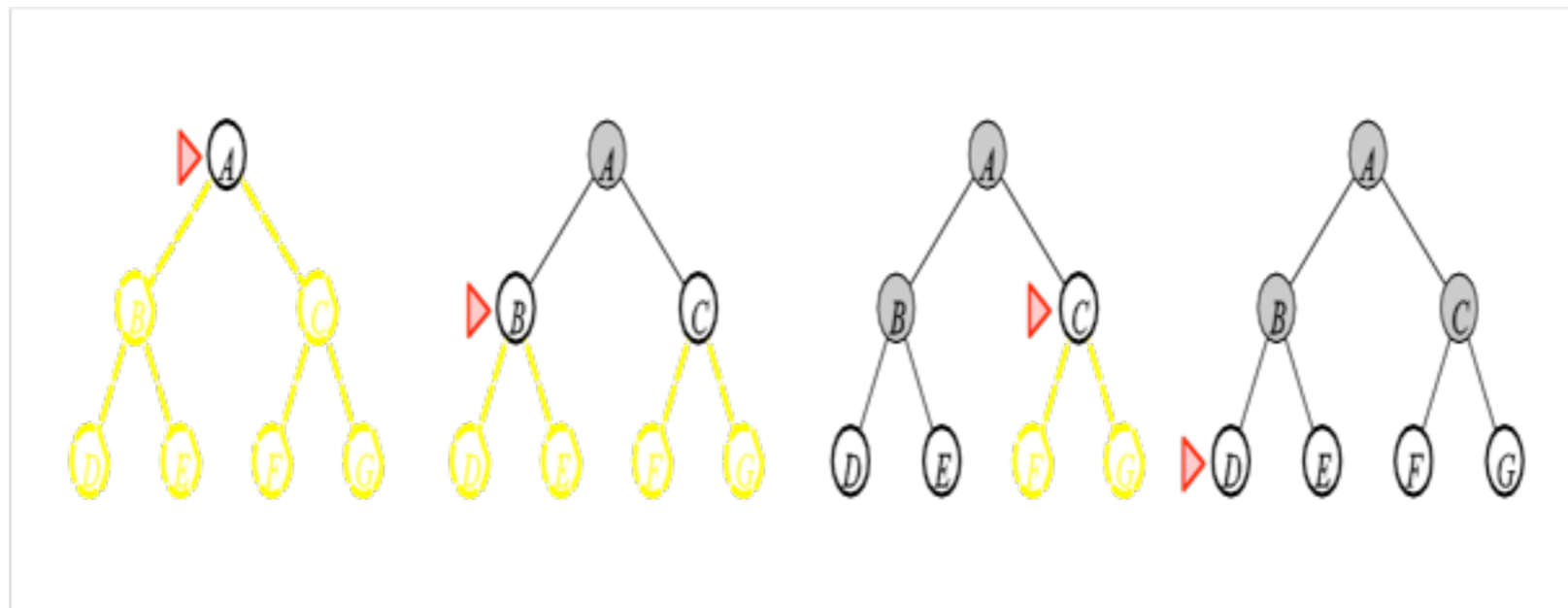
- Initialize with initial state of the problem
- Repeat
 - If no candidate nodes can be expanded **return failure**
 - Choose leaf node for expansion, according to **search strategy**
 - If node contains goal state, **return solution**
 - Otherwise, expand the node. Add resulting nodes to the tree

Implementation Details

- Need to keep track of nodes to be expanded (**fringe**)
- Implement using a queue:
 - Insert node for initial state
 - Repeat
 - If queue is empty, return failure
 - **Dequeue a node**
 - If node contains goal state, return solution
 - Expand node
- Search algorithms differ in their queuing function!

Breadth First Search

- All nodes on a given depth are expanded before any nodes on next level are expanded.
- Implemented with a **FIFO** queue



Key Properties

- **Completeness:** Is the alg. guaranteed to find a solution if the solution exists?
- **Optimality:** Does the alg. find the optimal solution?
- **Time complexity:** How many operations are needed?
- **Space complexity:** How much storage is needed?
- Other desirable properties
 - Can the alg. return an intermediate solution?
 - Can an adequate solution be refined or improved?

Search Performance

- Evaluated in terms of 2 characteristics
 - **Branching factor of state space:** how many operators can be applied at any time?
 - **Solution depth:** how long is the path to the closest solution?

b	Branching factor
d	Depth of shallowest goal node
m	Maximum length of any path in the state space

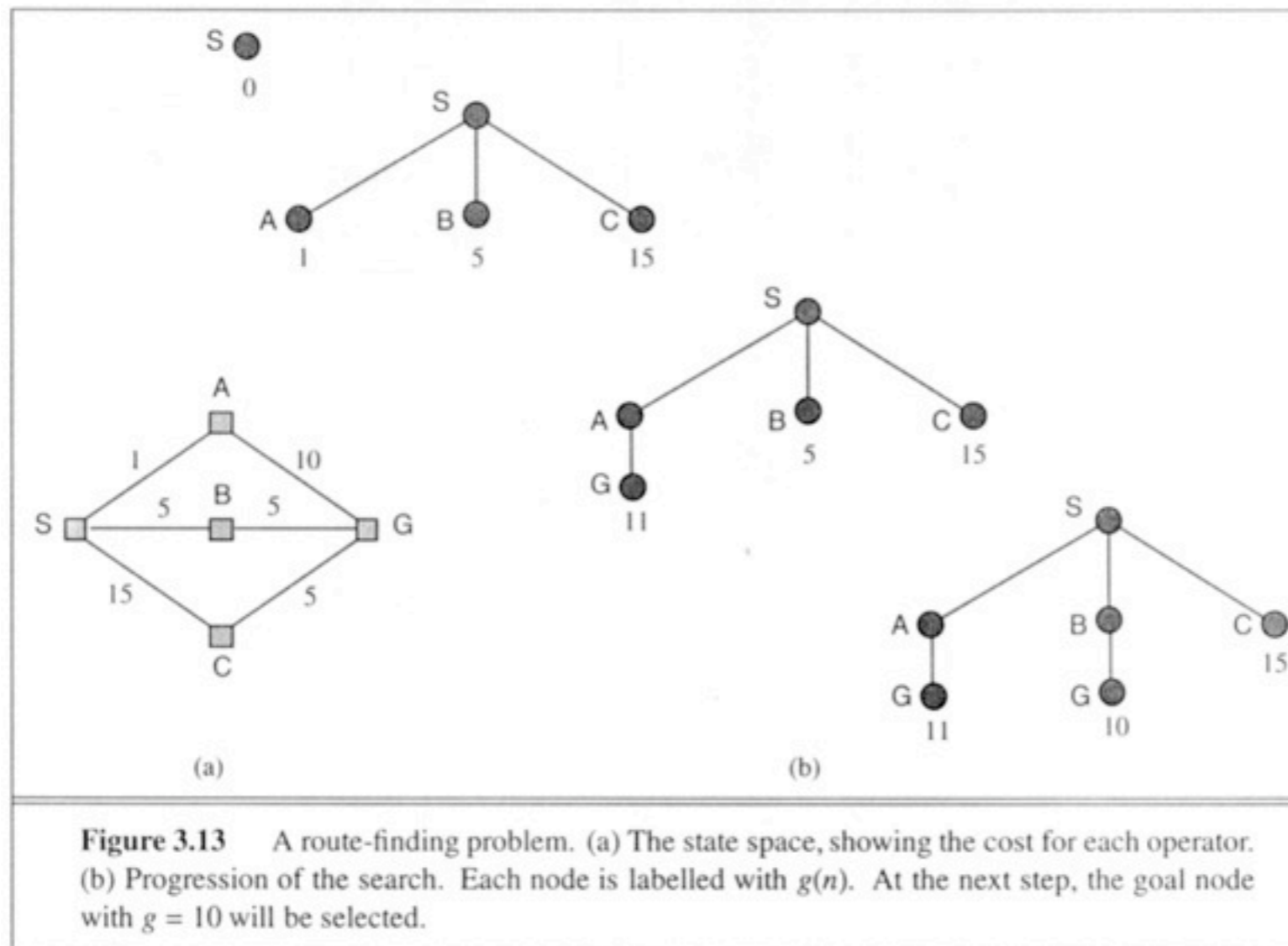
Judging BFS

- Good news
 - Complete (if b is finite)
 - Optimal (if all costs are the same)
- Bad news
 - Exponential time complexity: $O(b^{d+1})$
 - A problem with all uninformed search methods
 - Exponential space complexity: $O(b^{d+1})$
 - **Horrible!**

Making BFS Always Optimal

- Uniform cost search
 - Use a priority queue instead of a simple queue
 - Insert nodes in increasing order of the cost of path so far
 - Guaranteed to find optimal solution

Uniform Cost Search



C^* is cost of optimal solution

ϵ is minimum action cost

Time:

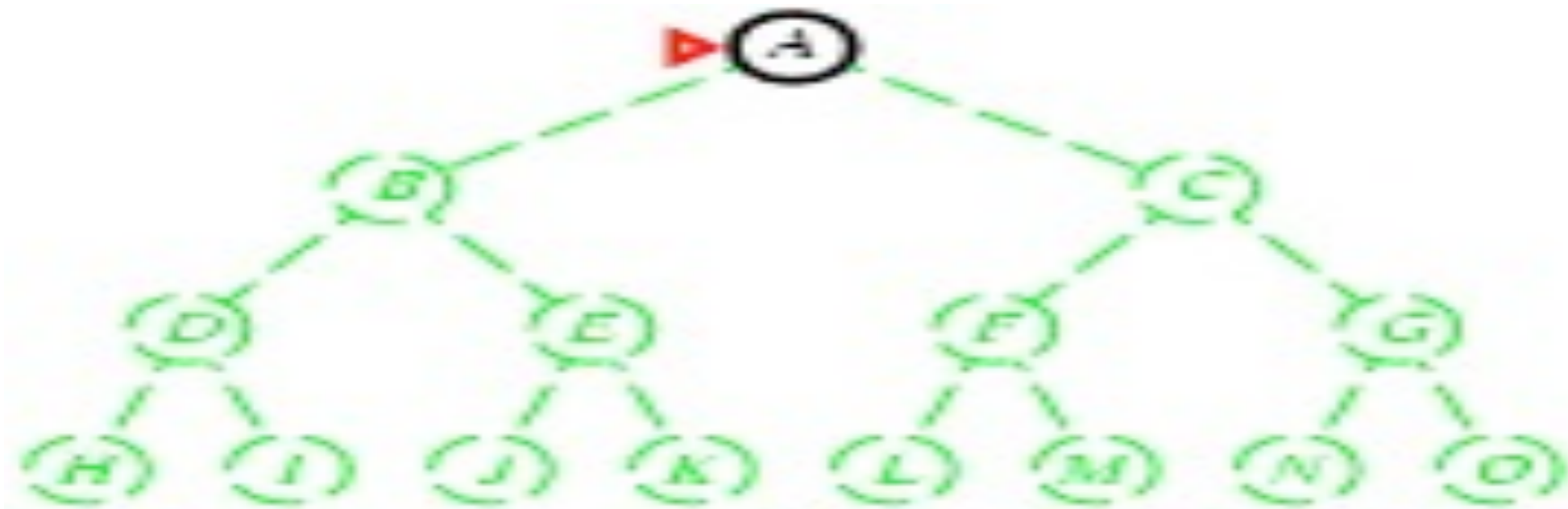
$O(b^{1+\text{floor}(C^*/\epsilon)})$

Space:

$O(b^{1+\text{floor}(C^*/\epsilon)})$

Depth Search Search

- Deepest node in current fringe of the search tree is expanded first
- Implemented with a stack



Judging Depth First Search

- Bad news
 - Not complete: might get stuck going down a long path
 - Not optimal: might return a solution at greater depth than another solution
 - Time complexity: $O(b^m)$
 - m might be much larger than d
- Good news
 - Space complexity: $O(bm)$

Depth Limited Search

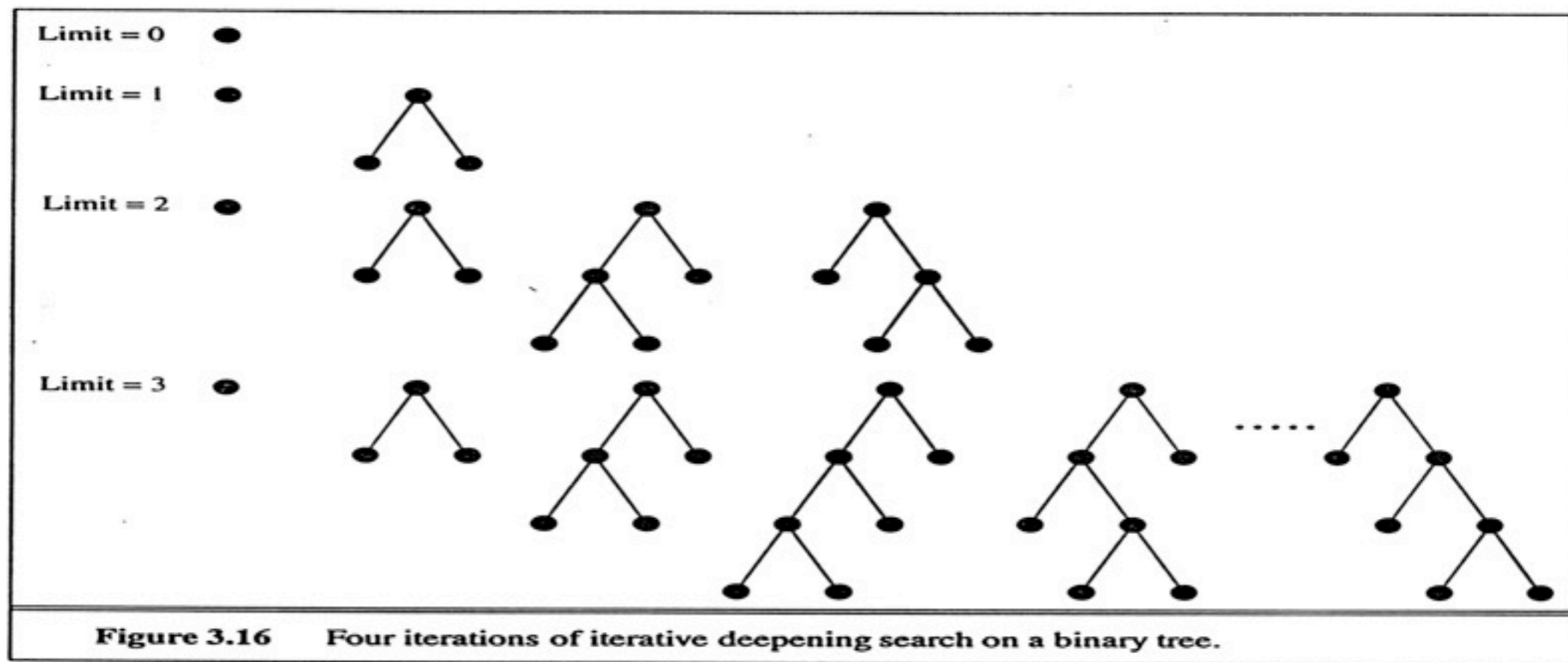
- Search depth-first, but terminate path if
 - a goal is found, or
 - maximum depth, l , is reached.
- How do you set l ?
- What happens is $l=1$?

Depth Limited Search

- Good news
 - Always terminates
 - Space: $O(b^l)$
- Bad news
 - Not complete
 - Goal depth might be deeper than l
 - Time: $O(b^l)$

Iterative Deepening

- Depth limited search, but increase the limit each iteration



Judging Iterative Deepening

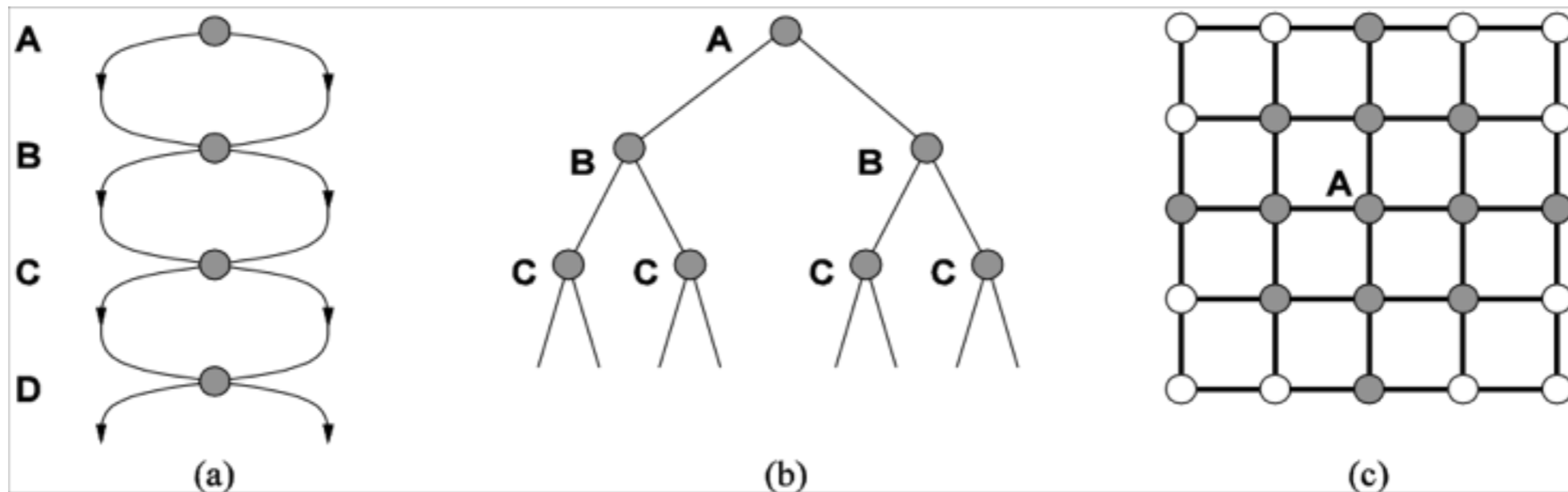
- Bad news
 - Time: $O(b^d)$
- Good news
 - Complete (like BFS)
 - Optimal
 - Space: **$O(bd)$**

Iterative Deepening

- Isn't IDS very wasteful?
 - Expanding the same nodes multiple times
- Insight
 - Most nodes are found in bottom level of the tree

Revisiting States

- What if we revisit a state that has already been expanded?
- What if we visit a state that is already in the queue?



Revisiting States

- Maintain a closed list to store every expanded node
 - More efficient on problems with many repeated states
 - Worst-case time and space are $O(S)$
 - S is number of states
- Allowing states to be re-expanded can produce a better solution
 - What should you do?

Summary

- Assumes no knowledge about the problem
 - General but expensive since we assume no knowledge about the problem
- Variety of uninformed search strategies
 - Mainly differ in the order in which they consider states

Criteria	BFS	Uniform	DFS	DLS	IDS
Complete	Yes	Yes	No	No	Yes
Time	$O(b^{d+1})$	$O(b^{1+\text{floor}(C^*/\epsilon)})$	$O(b^m)$	$O(b^l)$	$O(b^d)$
Space	$O(b^{d+1})$	$O(b^{1+\text{floor}(C^*/\epsilon)})$	$O(bm)$	$O(bl)$	$O(bd)$
Optimal	Yes	Yes	No	No	Yes

Questions?

- Next class: Informed search